

With this lemma a Lyapunov function $V = x^T P x$ can be constructed such that P satisfies not only the Lyapunov equation (D.15) but also the input-output condition (D.16) from which the restriction to relative degree zero ($d > 0$) or one ($d = 0, cb > 0$) is apparent. The main utility of this special Lyapunov function for adaptive and cascade designs is that the indefinite term in its derivative \dot{V} depends on the output y and not on the whole state x .

Appendix E

Parameter Projection

The modular adaptive controllers have a point of singularity $\hat{b}_m = 0$, where \hat{b}_m is the estimate of the high-frequency gain (virtual control coefficient) b_m . In order to prevent \hat{b}_m from taking the value zero, we use the parameter projection in our identifiers. For this, we need to know the sign of the actual high-frequency gain b_m . We first give a treatment of projection for a general convex parameter set and then specialize to the case where only the high-frequency gain is constrained.

Let us define the following convex set

$$\Pi = \{\hat{\theta} \in \mathbb{R}^p \mid \mathcal{P}(\hat{\theta}) \leq 0\}, \quad (\text{E.1})$$

where by assuming that the convex function $\mathcal{P} : \mathbb{R}^p \rightarrow \mathbb{R}$ is smooth, we assure that the boundary $\partial\Pi$ of Π is smooth. Let us denote the interior of Π by $\overset{\circ}{\Pi}$ and observe that $\nabla_{\hat{\theta}}\mathcal{P}$ represents an outward normal vector at $\hat{\theta} \in \partial\Pi$. The standard projection operator is

$$\text{Proj}\{\tau\} = \begin{cases} \tau, & \hat{\theta} \in \overset{\circ}{\Pi} \text{ or } \nabla_{\hat{\theta}}\mathcal{P}^T \tau \leq 0 \\ \left(I - \Gamma \frac{\nabla_{\hat{\theta}}\mathcal{P} \nabla_{\hat{\theta}}\mathcal{P}^T}{\nabla_{\hat{\theta}}\mathcal{P}^T \Gamma \nabla_{\hat{\theta}}\mathcal{P}}\right) \tau, & \hat{\theta} \in \partial\Pi \text{ and } \nabla_{\hat{\theta}}\mathcal{P}^T \tau > 0, \end{cases} \quad (\text{E.2})$$

where Γ belongs to the set \mathcal{G} of all positive definite symmetric $p \times p$ matrices. Although Proj is a function of three arguments, τ , $\hat{\theta}$ and Γ , for compactness of notation we write only $\text{Proj}\{\tau\}$.

The meaning of (E.2) is that, when $\hat{\theta}$ is in the interior of Π or at the boundary with τ pointing inward, then $\text{Proj}\{\tau\} = \tau$. When $\hat{\theta}$ is at the boundary with τ pointing outward, then Proj projects τ on the hyperplane tangent to $\partial\Pi$ at $\hat{\theta}$.

In general, the mapping (E.2) is discontinuous. This is undesirable for two reasons. First, the discontinuity represents a difficulty for implementation in continuous time. Second, since the Lipschitz continuity is violated, we cannot

where $i = 2, \dots, p$, and Γ_{1i} is the $(1, i)$ element of the positive definite symmetric matrix Γ . This update law achieves $\hat{b}_m(t) \operatorname{sgn} b_m \geq \varsigma_m - \varepsilon, \forall t \geq 0$ whenever $\hat{b}_m(0) \operatorname{sgn} b_m \geq \varsigma_m - \varepsilon$.

From expression (E.12) one should observe that when the update law is gradient it can be simplified to $\dot{\hat{\theta}}_i = \tau_i$ by the choice of the adaptation gain matrix $\Gamma = \operatorname{diag}\{\gamma_1, \Gamma_2\}$, where γ_1 is a positive scalar and Γ_2 is a symmetric positive definite $(p-1) \times (p-1)$ matrix. Thus, projection is applied only to $\dot{\hat{b}}_m$. This simplification is not possible with the least-squares update law because $\Gamma(t)$ is not a constant matrix, so it does not maintain a block-diagonal structure even if $\Gamma(0)$ is block-diagonal.

By setting $\varepsilon = 0$ in (E.11) and (E.12), we obtain update laws with the standard discontinuous projection, which in the case of gradient estimation simplifies to

$$\dot{\hat{b}}_m = \begin{cases} \tau_1 & \hat{b}_m \operatorname{sgn} b_m > \varsigma_m \text{ or } \tau_1 \operatorname{sgn} b_m \geq 0 \\ 0 & \hat{b}_m \operatorname{sgn} b_m \leq \varsigma_m \text{ and } \tau_1 \operatorname{sgn} b_m < 0 \end{cases} \quad (\text{E.13})$$

For the most part, the properties of the projection operator given in this appendix are recapitulated from [157] (see also [43, 51, 165]).

Appendix F

Nonlinear Swapping

The well-known Swapping Lemma [138] is ubiquitous in adaptive linear control. Here we provide its nonlinear counterpart.

Lemma F.1 (Nonlinear Swapping Lemma) *Consider the nonlinear time-varying system*

$$\begin{aligned} \Sigma_1 : \quad \dot{z} &= A(z, t)z + g(z, t)W(z, t)^T \tilde{\theta} - Q(z, t)^T \dot{\tilde{\theta}} + E(z, t)e \\ y_1 &= h(z, t)z + l(z, t)W(z, t)^T \tilde{\theta} \end{aligned} \quad (\text{F.1})$$

where $\tilde{\theta} : \mathbb{R}_+ \rightarrow \mathbb{R}^p$ is differentiable, $e : \mathbb{R}_+ \rightarrow \mathbb{R}$ is continuous and $\lim_{t \rightarrow \infty} e(t) = 0$, the matrix-valued functions $A : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{n \times n}$, $g : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{n \times m}$, $W : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{p \times m}$, $Q : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{p \times n}$, $E : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^n$, and $l : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{r \times m}$ are locally Lipschitz in z and continuous and bounded in t , and $h : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}^{r \times n}$ is bounded in z and t . Along with (F.1) consider the linear time-varying systems

$$\Sigma_2 : \quad \begin{aligned} \dot{\Omega}^T &= A(z, t)\Omega^T + g(z, t)W(z, t)^T \\ y_2 &= h(z, t)\Omega^T + l(z, t)W(z, t)^T \end{aligned} \quad (\text{F.2})$$

$$\Sigma_3 : \quad \begin{aligned} \dot{\psi} &= A(z, t)\psi + \Omega^T \dot{\tilde{\theta}} + Q(z, t)^T \dot{\tilde{\theta}} \\ y_3 &= -h(z, t)\psi. \end{aligned} \quad (\text{F.3})$$

Assume that $z(t)$ is continuous on $[0, \infty)$ and there exists a continuously differentiable function $V : \mathbb{R}^n \times \mathbb{R}_+ \rightarrow \mathbb{R}_+$ such that

$$\alpha_1 |\zeta|^2 \leq V(\zeta, t) \leq \alpha_2 |\zeta|^2, \quad (\text{F.4})$$

and for each $z \in C^0$,

$$\frac{\partial V}{\partial \zeta} A(z, t)\zeta + \frac{\partial V}{\partial t} \leq -\alpha_3 |\zeta|^2 - \alpha_4 \left(\frac{\partial V}{\partial \zeta} E(z, t) \right)^2 \quad (\text{F.5})$$